# **Increasing Object-Level Reconstruction Quality in Single-Image 3D Scene Reconstruction**

Anna Ribic

Antonio Oroz Meikel Kokowski Technical University of Munich Franz Srambical

{firstname}.{lastname}@tum.de

## **Abstract**

## 1. Introduction

While humans can easily infer the 3D structure as well as the complete (panoptic) semantics of a scene from a single image, this task has been a longstanding challenge in the field of computer vision. The task fundamentally prerequisites learning a strong prior of the 3D world. Traditional methods have made significant strides, from generating geometrically coherent structures [6, 19] to learning different instance semantics [7, 12, 18]. More recent approaches directly learn the 3D panoptic semantics as a whole [5, 28], yet they fall short in capturing the intricate details and nuances at the object level. This paper introduces a novel approach to bridge this gap by integrating a specialized object-level model into the reconstruction process, thereby leveraging the specialized model's object-priors.

# 2. Related Work

**2D panoptic segmentation** 2D panoptic segmentation merges semantic and instance segmentation, providing detailed pixel-level parsing of images, capturing both general categories (semantic segmentation) and individual object identities (instance segmentation) [10]. Since the original task formulation by Kirillov et al. [10], a number of works have been proposed to solve the task [1–3, 11, 13–15, 17, 21, 22, 25–27], while more recent approaches [9] try to unify image segmentation in its entirety.

**Single-view 3D reconstruction** The work by Snavely et al. [20] was the first notable attempt at reconstructing 3D scenes from unordered photo collections. Since then, the field of image-based 3D reconstruction has seen a number of advancements, culminating in the task of single-view 3D reconstruction [4, 6, 8, 16, 18, 19, 23].

**Shape priors** Wu et al. [24] note that the task of single-view 3D reconstruction is non-deterministic, as there are many 3D shapes that can explain a given single-view input, and propose to use shape priors to shape the solution space such that the reconstructed shapes are realistic, but not necessarily the ground truth.

# 3D scene understanding and panoptic reconstruction

#### DATASET

- 3. Method
- 4. Conclusion

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